

# Semi-automatic implementation of the complementary error function

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ARITH-26  
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- Mathematical functions are costly
  - rich trade-off possibilities
- Standard libm is not enough
- A "flavor" per application/target platform
  - high human resource consumption

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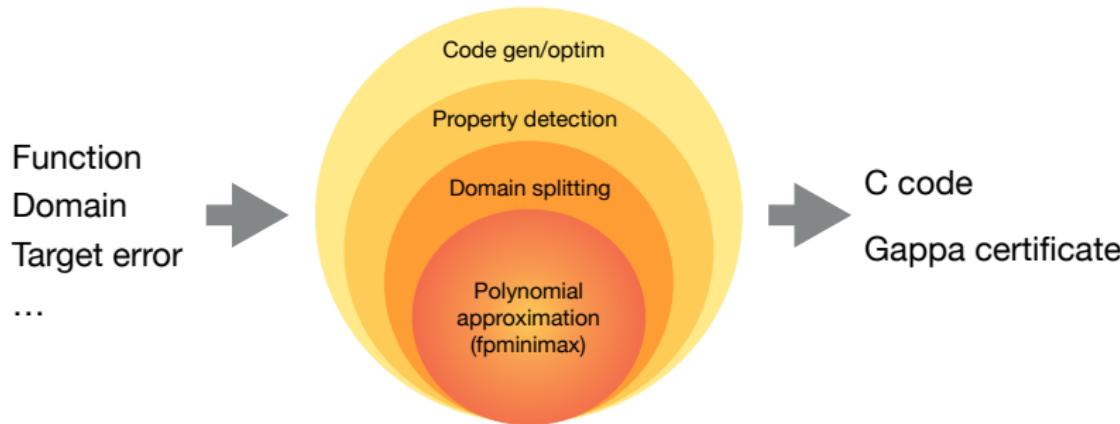
## Our approach:

- Automate
- Generate code on-demand
- Adapt for specific context



# Metalibm

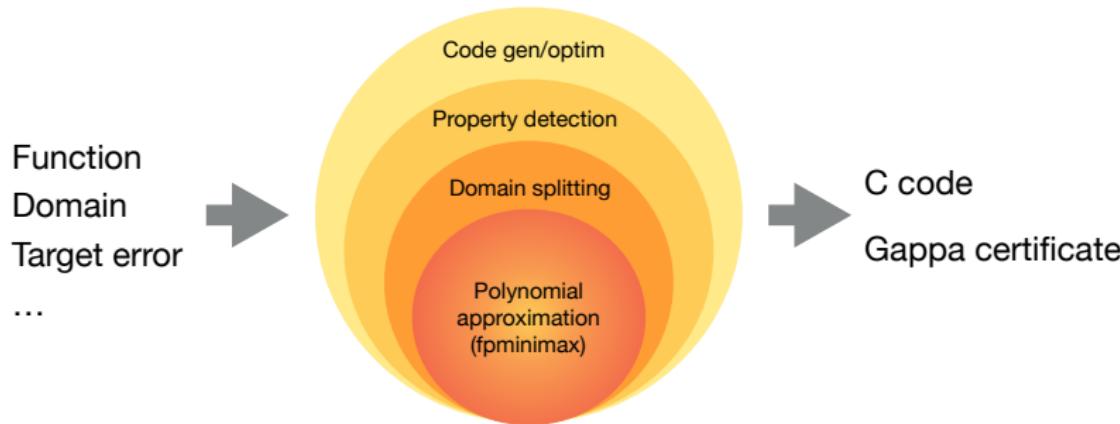
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- Easy to use
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# Metalibm

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- Deals with a variety of elementary functions ...**but special functions remain a challenge**

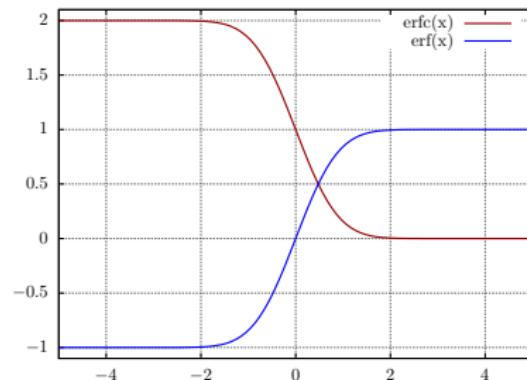
# Erf and erfc

$$\text{erf}(x) = \frac{2}{\sqrt{\pi}} \int_0^x e^{-t^2} dt$$

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Quick convergence:

- $\text{erf}(6) = 1$  in binary64
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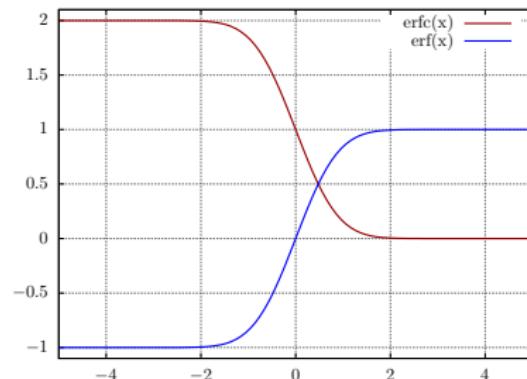
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## Metalibm with binary64 target accuracy:

- deals with  $\text{erf}(x)$  on  $[0; 6]$  within 49 sec
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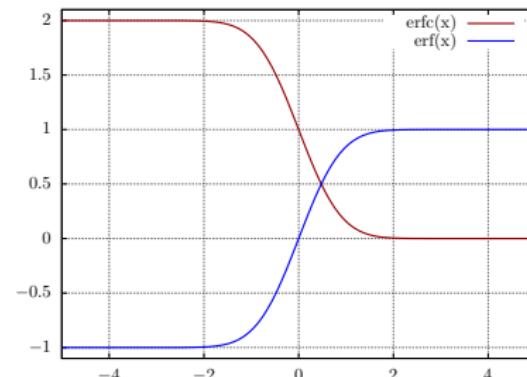
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## Reason:

- $\text{erfc}(x)$  is too "flat"
- not close enough to asymptotic expression  $e^{-x^2}/(x\sqrt{\pi})$

# Code generation for the $\text{erfc}(x)$

**Input:** relative error bound  $\delta$

**Output:** C code using binary64 data/arithmetic

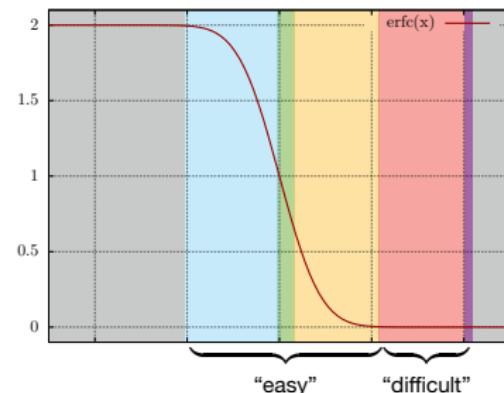
## Our approach:

“Easy” zones:

- directly use Metalibm

“Difficult” zone:

- asymptotic expression
- correction back to  $\text{erfc}(x)$
- re-partition of the error budget

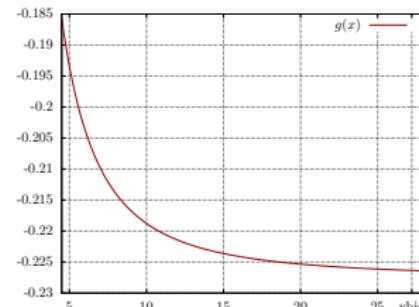


# Approximation technique

Easier-to-approximate function:

$$g(x) = \frac{1}{xe^{x^2} \operatorname{erfc}(x)} - 2$$

- decreasing on  $[5; x_{\text{BIG}}]$
- $|g(x)| \leq 2 - \sqrt{\pi} \leq 0.228$



Correction:

$$\operatorname{erfc}(x) = \frac{e^{-x^2}}{2x + xg(x)}$$

Evaluation:

- approximate exp and  $g$
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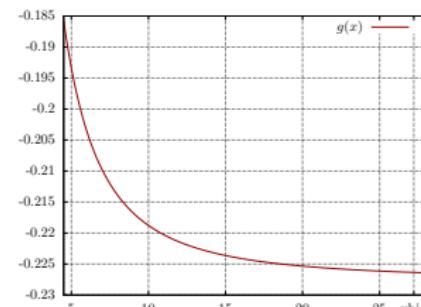
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Issue:

- $-x^2 \in [-741.256, -25]$
- exp will *underflow*

# What is the best way to scale?

Choose a scaling  $s$  to be within  $[-708.396 \dots; 670.96 \dots]$

$$e^{-x^2+s-s}$$

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Search for  $k \in \mathbb{Z}$  that minimizes  $|s - o(k \ln 2)|$

- for FP representation  $k = 61$ ,  
 $-x^2 + \hat{s} \in [-698.9 \dots, 17.2 \dots]$
- for DD representation  $k = 1021$ ,  
 $-x^2 + \hat{s} \in [-33.5 \dots, 682.7 \dots]$

# Error analysis and repartition

**Task:** ensure a relative error  $\delta$  and deduce accuracy of each step in

$$\text{erfc}(x) = 2^{-k} \frac{e^{-x^2 + \hat{s}}}{2x + xg(x)}$$

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$$y(x) = 2^{-k} a(x)/d(x)$$

$$a(x) = e^{t(x)}$$

$$t(x) = -x^2 + \hat{s}$$

$$d(x) = 2x + r(x)$$

$$r(x) = xg(x)$$

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$$\begin{aligned}\hat{a}(x) &= a(x)(1 + \varepsilon_a), \\ \hat{d}(x) &= d(x)/(1 + \varepsilon_d) \\ \hat{y}(x) &= 2^{-k} \frac{a(x)}{d(x)} (1 + \varepsilon_{\text{DIV}})(1 + \varepsilon_a)(1 + \varepsilon_d) \\ &= 2^{-k} \frac{a(x)}{d(x)} (1 + \varepsilon_y)\end{aligned}$$

To ensure  $\varepsilon_y \leq \varepsilon$  it suffices to ensure

$$|\varepsilon_a| \leq \varepsilon/4, \quad |\varepsilon_d| \leq \varepsilon/4, \quad |\varepsilon_{\text{DIV}}| \leq \varepsilon/4.$$

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$$\hat{t}(x) = t(x) + \Delta_t$$

$$\begin{aligned}\hat{a}(x) &= \text{EXP}(t(x) + \Delta_t) = e^{t(x)+\Delta_t} (1 + \varepsilon_{\text{EXP}}) \\&= e^{t(x)}(1 + e^{\Delta_t} - 1)(1 + \varepsilon_{\text{EXP}}) \\&= e^{t(x)} (1 + \varepsilon_a),\end{aligned}$$

To ensure  $\varepsilon_a \leq \varepsilon$  it suffices to ensure

$$|\varepsilon_{\text{EXP}}| \leq \varepsilon/4, \quad |\Delta_t| \leq \ln(1 + \varepsilon/4)$$

# Generic error bounds

Computation step	Error	Examples of error requirements		
	$\varepsilon_y$	$\delta$	$2^{-32}$	$2^{-46}$
$y(x) = 2^{-k} a(x)/d(x)$	$\varepsilon_{\text{DIV}}$	$\delta/4$	$2^{-34}$	$2^{-48}$
$a(x) = e^{t(x)}$	$\varepsilon_{\text{EXP}}$	$\delta/16$	$2^{-36}$	$2^{-50}$
$t(x) = -x^2 + k \ln 2$	$\Delta_t$	$\ln(1 + \delta/16)$	$1.99 \cdot 2^{-37}$	$1.99 \cdot 2^{-51}$
$d(x) = 2x + r(x)$	$\varepsilon_{\text{ADD}}$	$\delta/8$	$2^{-35}$	$2^{-49}$
$r(x) = xg(x)$	$\varepsilon_{\text{MUL}}$	$\frac{\delta}{4\bar{\alpha}(8+\delta)}$	$1.94 \cdot 2^{-35}$	$1.94 \cdot 2^{-49}$
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... but what happens in double precision?

# When straightforward binary64 is used

Computation step	Error	Bounds
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$y(x) = 2^{-k} a(x)/d(x)$	$ \varepsilon_{\text{DIV}} $	$u$
$a(x) = e^{t(x)}$	$ \varepsilon_{\text{EXP}} $	$1 \dots \delta - 1024.2584u$
$t(x) = -x^2 + k \ln 2$	$ \Delta_t $	$1024.2583u$
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Must be more accurate in critical parts

# Exploiting double-word arithmetic

- Evaluate  $t(x)$  as a double-word  $t_h + t_\ell$

## Method 1

$$|\Delta_t| \leq 32.259u$$

6 FP operations

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$$e^{t_h}(1 + t_\ell)(1 + \varepsilon_{E_\ell})(1 + \varepsilon_t)$$

$$|\varepsilon_t| \leq 0.2585u$$

$$|\varepsilon_{E_\ell}| \leq (1089 \cdot 2^{44})u$$

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$$\begin{cases} |\varepsilon_t| \leq 0.2585u \\ |\varepsilon_{E_\ell}| \leq (1089 \cdot 2^{44})u \\ |\varepsilon_{\text{FMA}}| \leq u \end{cases}$$

**Result:** can approximate with error up to  $\delta = 0.76 \cdot 2^{-50}$   
**Cost:** 13 FP operations

# Numerical results – 1

## Implementation:

- Semi-automatic approximation choice for Metalibm
- Code generation in C

## Testing:

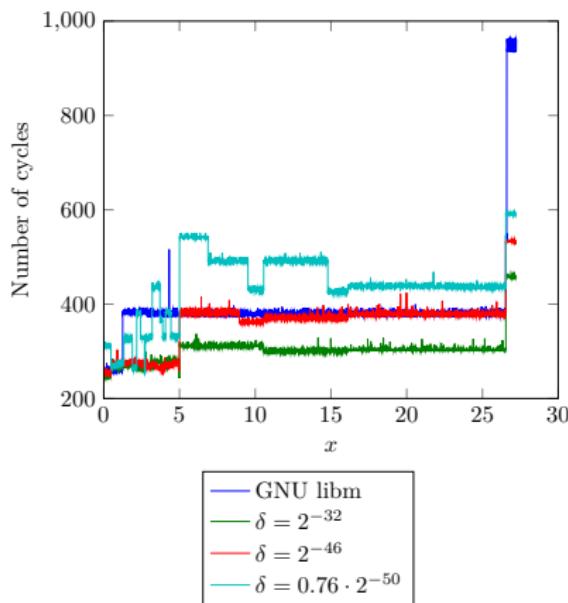
- Reference implementation: GNU libm with gcc 6.3.0
- Target accuracy:  $2^{-32}, 2^{-46}, 0.76 \cdot 2^{-50}$

## Results:

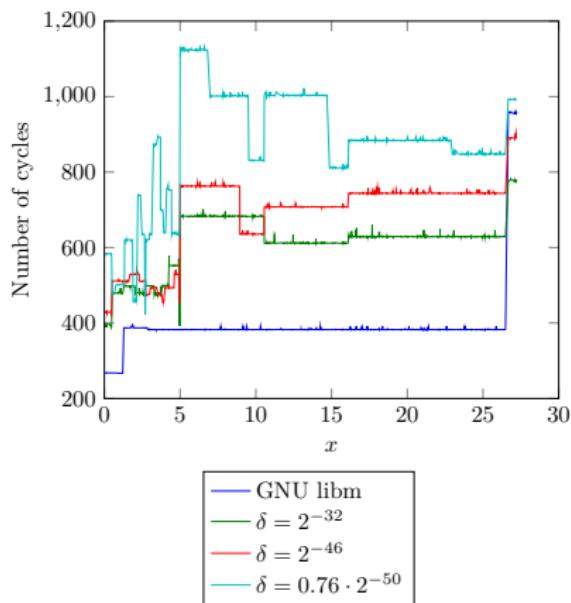
accuracy	[0; 5]		[5; $x_{\text{LARGE}}$ ]		$[x_{\text{LARGE}}, x_{\text{BIG}}]$ abs
	abs	rel	abs	rel	
GNU libm	4 ulp	6.34 u	3 ulp	3.98 u	1.5 ulp
$0.76 \cdot 2^{-50}$	2 ulp	3.84 u	4 ulp	4.02 u	1.5 ulp
$2^{-46}$	18 ulp	21.07 u	15 ulp	16.6 u	1.5 ulp

# Numerical results – 2

-03



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# Conclusion

- Partly-automated implementation that offers
  - a priori target accuracy
  - guaranteed error bounds
  - exploration of a large design space
- Asymptotic expression + correction
- Double-word arithmetic for critical parts when in binary64

# Perspectives

- Optimize error budget repartition
- Achieve higher accuracy
- Adapt for other functions with similar behavior, e.g. Gaussian